Package ‘scMappR’

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Title Single Cell Mapper

Version 0.1.3

Description
Statistical tools to interrogate the cell-type specificity of any gene list given a matrix of cell-types and genes associated with those cell-types (a signature matrix). Additionally, a library of over 245 signature matrices from public data are stored. Together, there are four primary functions: (i) processing scRNA-seq count data and automated cell-type naming using 'Seurat' V3 and enrichment of CellMarker and Panglao databases, (ii) tissue-by-cell-type gene set enrichment, (iii) cell-type specific enrichment of a gene list within a particular tissue, (iv) weighted cell-type specific reranking of a list of differentially expressed genes. Reference: Sokolowski DJ, Faykoo-Martinez M, Erdman L, Hou H, Uuskula-Reimand L, Yuki K, Zhu H, Goldenberg A, Wilson MD (In Preparation).

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`cellmarker_enrich` *Fisher's Exact Cell-Type Identification.*

Description

This function uses the CellMarker and Panglao datasets to identify cell-type differentially expressed genes.

Usage

```
cellmarker_enrich(
  gene_list,
  p_thresh,
  gmt = "cellmarker_list.Rdata",
  fixed_length = 13000,
  min_genes = 5,
  max_genes = 3000,
  isect_size = 3
)
```

Arguments

`gene_list`	A character vector of gene symbols with the same designation (e.g. mouse symbol - mouse, human symbol - human) as the gene set database.
`p_thresh`	The Fisher's test cutoff for a cell-marker to be enriched.
`gmt`	Either a path to an rda file containing an object called "gmt", which is a named list where each element of the list is a vector of gene symbols website for more detail on the file type (https://software.broadinstitute.org/cancer/software/gsea/wiki/index.php/Data_form The gmt list may also be inputted.
`fixed_length`	Estimated number of genes in your background.
`min_genes`	Minimum number of genes in the cell-type markers.
`max_genes`	Maximum number of genes in the cell-type markers.
`isect_size`	Number of genes in your list and the cell-type.

Details

Complete a Fisher's exact test of an input list of genes against a gene set saved in an *.RData object. The RData object is storing a named list of genes called "gmt".

Value

`cellmarker_enrich` Gene set enrichment of cell-types on your inputted gene list.

Examples

```
data(POA_example)
POA_generes <- POA_example$POA_generes
POA_OR_signature <- POA_example$POA_OR_signature
POA_Rank_signature <- POA_example$POA_Rank_signature
Signature <- POA_Rank_signature
rowname <- get_gene_symbol(Signature)
rownames(Signature) <- rowname$rowname
genes <- rownames(Signature)[1:100]
data(gmt)
```

```
enriched <- cellmarker_enrich(gene_list = genes, p_thresh = 0.05, gmt = gmt)
```

---

coEnrich *Identify co-expressed cell-types*

Description

This function identifies genes with similar cell-type markers and if those markers are driving enrichment.

Usage

```
coEnrich(
  sig,
  gene_list_heatmap,
  background_heatmap,
  study_name,
  outDir,
  toSave = FALSE,
  path = NULL
)
```

Arguments

`sig`	A The number of combinations of significant cell-types to enrich.
`gene_list_heatmap`	Signature matrix of inputted genes in heatmap and the cell-type preferences – output of heatmap generation.
`background_heatmap`	Signature matrix of background matrix in heatmap and cell-type preferences – output of heatmap generation.
`study_name`	Name of the outputted table.
`outDir`	Name of the directory this table will be printed in.
`toSave`	Allow scMappR to write files in the current directory (T/F).
`path`	If toSave == TRUE, path to the directory where files will be saved.

Details

This function takes significantly enriched cell-types from the single CT_enrich before testing to see if the genes driving their enrichment are overlapping to a significant proportion using Fisher's exact test. To save computational time and to not complete this with an incredible number of permutations, scMappR stops at overlapping 5 cell-types.

Value

coEnrich Enrichment of cell-types that are expressed by the same genes, up to 4 sets of cell-types.

Examples

```
# load in signature matrices
data(POA_example)
POA_generes <- POA_example$POA_generes
POA_OR_signature <- POA_example$POA_OR_signature
POA_Rank_signature <- POA_example$POA_Rank_signature
sig <- get_gene_symbol(POA_Rank_signature)
Signature <- POA_Rank_signature
rownames(Signature) <- sig$rowname
genes <- rownames(Signature)[1:60]
heatmap_test <- tissue_scMappR_custom( genes, signature_matrix = Signature,
output_directory =  "scMappR_test", toSave = FALSE)
group_preferences <- heatmap_test$group_celltype_preferences
```

DeconRNAseq_CRAN   *DeconRNASeq CRAN compatible*

Description

This function runs DeconRNAseq with default parameters such that it is compatible with CRAN and scMappR

Usage

```
DeconRNAseq_CRAN(
  datasets,
  signatures,
  proportions = NULL,
  checksig = FALSE,
  known.prop = FALSE,
  use.scale = TRUE,
  fig = FALSE
)
```

Arguments

datasets	Normalized RNA-seq dataset
signatures	Signature matrix of odds ratios
proportions	If cell-type proportion is already inputted - always NULL for scMappR
checksig	Check to see if plotting is significant - always false for scMappR

known.prop	If proportions were known - always false for scMappR
use.scale	Scale and center value - always TRUE for scMappR
fig	Make figures - always FALSE for scMappR

Details

This is the exact same function as the primary function in the Bioconductor package, DeconRNAseq (PMID: 23428642) except it is now compatible with CRAN packages.

Value

DeconRNAseq_CRAN Estimated cell-type proportions with DeconRNAseq.

Examples

```
data(PBMC_example)
bulk_DE_cors <- PBMC_example$bulk_DE_cors
bulk_normalized <- PBMC_example$bulk_normalized
odds_ratio_in <- PBMC_example$odds_ratio_in
out <- DeconRNAseq_CRAN(as.data.frame(bulk_normalized), as.data.frame(odds_ratio_in))
```

---

deconvolute_and_contextualize

Generate cell weighted Fold-Changes (cwFold-changes)

Description

This function takes a count matrix, signature matrix, and differentially expressed genes (DEGs) before generating cwFold-chagnes for each cell-type.

Usage

```
deconvolute_and_contextualize(
  count_file,
  signature_matrix,
  DEG_list,
  case_grep,
  control_grep,
  max_proportion_change = -9,
  print_plots = T,
  plot_names = "scMappR",
  theSpecies = "human",
  make_scale = FALSE,
  FC_coef = T,
```

```
    sig_matrix_size = 3000,
    sig_distort = 1,
    drop_unknown_celltype = TRUE,
    toSave = FALSE,
    path = NULL
  )
```

Arguments

`count_file` Normalized deconvolute_and_contextualize. RNA-seq count matrix where rows are gene symbols and columns are individuals. Either the object tself of the path of a .tsv file.

`signature_matrix` Signature matrix (recommended odds ratios) of cell-type specificity of genes. Either the object itself or a pathway to a .RData file containing an object named "wilcoxon_rank_mat_or" - generally internal.

`DEG_list` An object with the first column as gene symbols within the bulk dataset (doesn't have to be in signature matrix), second column is the adjusted P-value, and the third the log2FC. Path to a tsv file containing this info is also acceptable.

`case_grep` Tag in the column name for cases (i.e. samples representing upregulated) OR an index of cases.

`control_grep` Tag in the column name for control (i.e. samples representing downregulated) OR an index of cases.

`max_proportion_change` Maximum cell-type proportion change. May be useful if there are many rare cell-types.

`print_plots` Whether boxplots of the estimated CT proportion for the leave-one-out method of CT deconvolution should be printed (T/F).

`plot_names` If plots are being printed, the pre-fix of their .pdf files.

`theSpecies` -9 if using a precomputed count matrix from scMappR, human otherwise. Removes ensembl symbols if appended.

`make_scale` Convert the lowest odds ratio to 1 and scales accordingly – strongly not recommended and will produce warning if used.

`FC_coef` Making cwFold-changes based on fold-change (TRUE) or rank := (-log10(Pval)) (FALSE) rank. After testing, we strongly recommend to keep true (T/F).

`sig_matrix_size` Number of genes in signature matrix for cell-type deconvolution.

`sig_distort` Exponential change of odds ratios. Strongly not recomended and will produce warnings if changed from default.

`drop_unknown_celltype` Whether or not to remove "unknown" cell-types from the signature matrix (T/F).

`toSave` Allow scMappR to write files in the current directory (T/F).

`path` If toSave == TRUE, path to the directory where files will be saved.

Details

This function completes the cell-type contextualization in scMappR – reranking every DEG based on their fold change, likelihood the gene is in each detected cell-type, average cell-type proportion, and ratio of cell-type proportion between case and control. If a gene is upregulated then it is being controlled by control/case, otherwise it is case/control. cwFold-change's are generated for genes that are in both the count matrix and in the list of DEGs. It does not have to also be in the signature matrix. First, this function will estimate cell-type proportions with all genes included before estimating changes in cell-type proportion between case/control using a t-test. Then, it takes a leave-one-out approach to cell-type deconvolution such that estimated cell-type proportions are computed for every inputted DEG. Optionally, the differences between cell-type proprtion before and after a gene is removed is plotted in boxplots. Then, for every gene, cwFold-change's are computed with the following formula (the example for upreguatled genes) val <- cell-preferences * cell-type_proprtion * cell-type_proportion_fold-change * sign*2^abs(gene_DE$log2fc). A matrix of cwFold-change's for all DEGs are returned.

Value

List with the following elements:

`cellWeighted_Foldchange`
: data frame of cellweightedFold changes for each gene.

`cellType_Proportions`
: data frame of cell-type proportions from DeconRNA-seq.

`leave_one_out_proportions`
: data frame of average cell-type proportions for case and control when gene is removed.

`processed_signature_matrix`
: signature matrix used in final analysis.

Examples

```
data(PBMC_example)
bulk_DE_cors <- PBMC_example$bulk_DE_cors
bulk_normalized <- PBMC_example$bulk_normalized
odds_ratio_in <- PBMC_example$odds_ratio_in
case_grep <- "_female"
control_grep <- "_male"
max_proportion_change <- 10
print_plots <- FALSE
theSpecies <- "human"
norm <- deconvolute_and_contextualize(bulk_normalized, odds_ratio_in, bulk_DE_cors,
                                    case_grep = case_grep, control_grep = control_grep,
                                     max_proportion_change = max_proportion_change,
                                      print_plots = print_plots,
                                     theSpecies = theSpecies, toSave = FALSE)
```

extract_genes_cell *Extract Markers*

Description

Extracting cell-type markers from a signature matrix.

Usage

```
extract_genes_cell(
  geneHeat,
  cellTypes = "ALL",
  val = 1,
  isMax = FALSE,
  isPvalue = FALSE
)
```

Arguments

geneHeat	The heatmap of ranks from your scRNA-seq dataset with your genes subsetted.
cellTypes	The cell-types that you're interested in extracting. They need to be colnames (not case sensitive).
val	How associated a gene is with a particualr cell type to include in your list - default is slightly associated.
isMax	If you are taking the single best CT marker (T/F) – TRUE not recommended.
isPvalue	If the signature matrix is raw p-value (T/F) – TRUE not recommended.

Details

This function takes a signature matrix and extracts cell-type markers above a p-value or fold-change threshold.

Value

`extract_genes_cell` A vector of genes above the threshold for each sample.

Examples

```
data(POA_example)
Signature <- POA_example$POA_Rank_signature
RowName <- get_gene_symbol(Signature)
rownames(Signature) <-RowName$rowname
# extract genes with a -log10(Padj > 1)
Signat <- extract_genes_cell(Signature)
```

generes_to_heatmap   *Generate signature matrix*

Description

Convert a list of cell-type markers from FindMarkers in Seurat to a signature matrix defined by odds ratio and rank.

Usage

```
generes_to_heatmap(
  generes,
  species = "human",
  naming_preference = -9,
  rda_path = "",
  make_names = TRUE,
  internal = FALSE
)
```

Arguments

- `generes` A list of cell-tpe markers with fold-changes and p-vlaues (FindMarkers output in Seurat).
- `species` The species of gene symbols, if not internal, "human" or "mouse".
- `naming_preference` Likely cell-types given tissues (to be passed into human_mouse_ct_marker_enrich).
- `rda_path` Path to output direcotry, if toSave is true.
- `make_names` Identify names of cell-type markers using the Fisher's exact test method (T/F).
- `internal` If this function is pre-processing from Panglao (T/F).

Details

Take a list of compiled differentially expressed genes from different cell-types, identify what the cell-types are using the Fisher's exact test, and then convert into a signature matrix for both the adjusted p-value and odds ratio.

Value

List with the following elements:

- `pVal` A dataframe containing the signature matrix of ranks (-log10(Padj) * sign(fold-change)).

`OR`	A dataframe containing the signature matrix of odds ratios.
`cellname`	A vector of the cell-labels returned from the GSVA method.
`topGenes`	the top 30 mos expressed genes in each cell-type.

Examples

```
data(POA_example)
 POA_generes <- POA_example$POA_generes
signature <- generes_to_heatmap(POA_generes,species = -9, make_names = FALSE)
```

`get_gene_symbol` *Internal – get gene symbol from Panglao.db matrix.*

Description

Internal – removes Ensembl signature appended to signature matrix from Panglao and figure out species by pre-fix Ensembl of the Ensembl ID that is appended to gene names.

Usage

```
get_gene_symbol(wilcoxon_rank_mat_t)
```

Arguments

`wilcoxon_rank_mat_t`
Matrix where row names are "GeneSymbol-Ensembl" (human or mouse).

Details

Internal: This function removes the ENGMUS/ENGS tag from Panglao created gene names (symbol-ENGS). From the ENSG/ENSMUS, this function determines if the species is mouse/human and returns the gene symbols.

Value

List with the following elements:

`rowname`	Genes in the signature matrix excluding the ensemble name.
`species`	"mouse" or "human" depending on appended ensembl symbols.

Examples

```
 # load signature
data(POA_example)
POA_OR_signature <- POA_example$POA_OR_signature
symbols <- get_gene_symbol(POA_OR_signature)
```

---

gmt *gmt_example*

Description

Markers of 5 glial cell-types

Usage

```
data(gmt)
```

Format

A list with 5 character vectors, each containing genes.

Astrocytes_panglao astrocyte markers identified by panglao

Schwann_panglao Schwann markers identified by panglao

Bergmann glia_panglao Bergmann glia markers identified by panglao

Kupffer_panglao Kupffer markers identified by panglao

Oligodendrocyte progenitor_panglao Oligodendrocyte progenitor markers identified by panglao

Details

A named list containing the cell-type markers of 5 glial cell types. Used for testing cell-type naming functions.

Examples

```
data(gmt)
```

gProfiler_cellWeighted_Foldchange
Pathway enrichment for cwFold-changes

Description

This function runs through each list of cell weighted Fold changes (cwFold-changes) and completes both pathway and transcription factor (TF) enrichment.

Usage

```
gProfiler_cellWeighted_Foldchange(
  cellWeighted_Foldchange_matrix,
  species,
  background,
  gene_cut,
  newGprofiler
)
```

Arguments

cellWeighted_Foldchange_matrix	Matrix of cell weighted Fold changes from the deconvolute_and_contextualize functions.
species	Human, mouse, or a charcter that is compatible with gProfileR.
background	A list of background genes to test against.
gene_cut	The top number of genes in pathway analysis.
newGprofiler	Using gProfileR or gprofiler2, (T/F).

Details

This function takes a matrix of cellWeighted_Foldchange and a species (human, mouse, or a character directly compatible with g:ProfileR). Before completing pathway analysis with g:ProfileR. Enriched pathways are stored in a list and returned.

Value

List with the following elements:

BP	gprofiler enrichment of biological pathways for each cell-type
TF	gprofiler enrichment of transcription factors for eachc cell-type.

Examples

```
data(PBMC_example)

bulk_DE_cors <- PBMC_example$bulk_DE_cors
bulk_normalized <- PBMC_example$bulk_normalized
odds_ratio_in <- PBMC_example$odds_ratio_in

case_grep <- "_female"
control_grep <- "_male"
max_proportion_change <- 10
print_plots <- FALSE
theSpecies <- "human"
norm <- deconvolute_and_contextualize(bulk_normalized, odds_ratio_in,
                                      bulk_DE_cors, case_grep = case_grep,
                                      control_grep = control_grep,
                                       max_proportion_change = max_proportion_change,
                                       print_plots = print_plots,
                                       theSpecies = theSpecies)
background = rownames(bulk_normalized)
STVs <- gProfiler_cellWeighted_Foldchange(norm$cellWeighted_Foldchange, theSpecies,
 background, gene_cut = -9, newGprofiler = FALSE)
```

`gsva_cellIdentify` *Cell-type naming with GSVA*

Description

This function computes the mean expression of every cell-type before predicting the most likely cell-type using the GSVA method.

Usage

```
gsva_cellIdentify(
  pbmc,
  theSpecies,
  naming_preference = -9,
  rda_path = "",
  toSave = FALSE
)
```

Arguments

pbmc Processed seurat object without named cells.

theSpecies "human" or "mouse" – it will determine which CT marker database to use – there are some differences.

naming_preference Once top CT markers are identified, naming_preferences will then extract CT markers within a more appropriate tissue type.

rda_path Path to pre-computed cell-type .gmt files (rda objects).

toSave If scMappR is allowed to write files and directories.

Details

This function inputs a Seurat object and uses the average normalized expression of each gene in each cluster to identify cell-types using the GSVA method.

Value

List with the following elements:

cellMarker Most likely cell-types predicted from cellMarker database.

panglao Most likely cell-types predicted from panglao database.

avg_expression Average expression of each gene in each cell-type.

Examples

```
data(sm)
toProcess <- list(example = sm)
tst1 <- process_from_count(toProcess, "testProcess")
cellnames <- gsva_cellIdentify(tst1, theSpecies = "mouse",
 naming_preference = "brain", rda_path = "")
```

---

heatmap_generation *Generate Heatmap*

Description

This function takes an inputted signature matrix as well as a list of genes and overlaps them. Then, if there is overlap, it prints a heatmap or barplot (depending on the number of overlapping genes). Then, for every cell-type, genes considered over-represented are saved in a list.

Usage

```
heatmap_generation(
  genesIn,
  comp,
  reference,
  cex = 0.8,
  rd_path = "",
  cellTypes = "ALL",
  pVal = 0.01,
  isPval = TRUE,
  isMax = FALSE,
  isBackground = FALSE,
  which_species = "human",
  toSave = FALSE,
  path = NULL
)
```

Arguments

genesIn	A list of gene symbols (all caps) to have their cell type enrichment.
comp	The name of the comparison.
reference	Path to signature matrix or the signature matrix itself.
cex	The size of the genes in the column label for the heatmap.
rd_path	The directory to RData files – if they are not in this directory, then the files will be downloaded.
cellTypes	Colnames of the cell-types you will extract (passed to extract_genes_cell).
pVal	The level of association a gene has within a cell type (passed to extract_genes_cell).
isPval	If the signature matrix is raw p-value (T/F) – TRUE not reccomended.
isMax	If you are taking the single best CT marker (T/F) – TRUE not reccomended.
isBackground	If the heatmap is from the entire signature matrix or just the inputted gene list (T/F). isBackground == TRUE is used for internal.
which_species	Species of gene symbols – "human" or "mouse" .
toSave	Allow scMappR to write files in the path directory (T/F).
path	If toSave == TRUE, path to the directory where files will be saved.

Value

List with the following elements:

genesIn	Vector of genes intersecting gene list and signature matrix.
genesNoIn	Vector of inputted genes not in signature matrix.
geneHeat	Signature matrix subsetted by inputted gene list
preferences	Cell-markers mapping to cell-types.

Examples

```
# load in signature matrices
data(POA_example)
POA_generes <- POA_example$POA_generes
POA_OR_signature <- POA_example$POA_OR_signature
POA_Rank_signature <- POA_example$POA_Rank_signature
Signature <- POA_Rank_signature
rowname <- get_gene_symbol(Signature)
rownames(Signature) <- rowname$rowname
genes <- rownames(Signature)[1:100]
heatmap_test <- heatmap_generation(genesIn = genes, "scMappR_test",
                                   reference = Signature, which_species = "mouse")
```

---

human_mouse_ct_marker_enrich
Consensus cell-type naming (Fisher's Exact)

Description

This function completes the Fisher's exact test cell-type naming for all cell-types.

Usage

```
human_mouse_ct_marker_enrich(
  gene_lists,
  theSpecies = "human",
  cell_marker_path = "",
  naming_preference = -9
)
```

Arguments

`gene_lists` A named list of vectors containing cell-type markers (mouse or human gene-symbols) which will be named as a cell-type via the Fisher's exact test method.

`theSpecies` The species of the gene symbols: "human" or "mouse".

`cell_marker_path` If local, path to cell-type marker rda files, otherwise, we will try to download data files.

`naming_preference` Either -9 if there is no expected cell-type or one of the categories from get_naming_preference_options(). This is useful if you previously have an idea of which cell-type you were going to enrich.

Details

Fisher's exact test method of cell-type identification using the Panglao and CellMarker databases. It extracts significant pathways (pFDR < 0.05). Then, if naming_preference != -9, it will extract the enriched cell-types within the cell-types identified with the naming preferences option. Generally, this method seems to be biased to cell-types with a greater number of markers.

Value

List with the following elements:

`cellTypes`	most likely marker for each cell-type from each database.
`marker_sets`	all enriched cell-types for each cluster from each dataset.

Examples

```
data(POA_example)
POA_generes <- POA_example$POA_generes
POA_OR_signature <- POA_example$POA_OR_signature
POA_Rank_signature <- POA_example$POA_Rank_signature
Signature <- POA_Rank_signature
rowname <- get_gene_symbol(Signature)
rownames(Signature) <- rowname$rowname
genes <- rownames(Signature)[1:100]
enriched <- human_mouse_ct_marker_enrich(gene_lists = genes, theSpecies = "mouse",
                                     cell_marker_path = "", naming_preference = "brain")
```

---

make_TF_barplot *Plot gprofileR Barplot (TF)*

Description

Make a barplot of the top transcription factors enriched by gprofileR.

Usage

```
make_TF_barplot(ordered_back_all_tf, top_tf = 5)
```

Arguments

`ordered_back_all_tf`	Output of the gprofileR function.
`top_tf`	The number of pathways you want to plot.

Details

This function takes a gprofileR output and prints the top "top_tfs" most significantly enriched p-values before plotting the rank of their p-values.

Value

`make_TF_barplot` A barplot of the number of "top_tf" tf names (not motifs), ranked by -log10(Pfdr).

Examples

```
data(POA_example)
 POA_generes <- POA_example$POA_generes
 POA_OR_signature <- POA_example$POA_OR_signature
 POA_Rank_signature <- POA_example$POA_Rank_signature
Signature <- as.data.frame(POA_Rank_signature)
rowname <- get_gene_symbol(Signature)
rownames(Signature) <- rowname$rowname
ordered_back_all <- gprofiler2::gost(query = rowname$rowname[1:100], organism = "mmusculus",
ordered_query = TRUE, significant = TRUE, exclude_iea = FALSE, multi_query = FALSE,
measure_underrepresentation = FALSE, evcodes = FALSE, user_threshold = 0.05,
correction_method = "fdr",   numeric_ns = "", sources = c("GO:BP", "KEGG", "REAC"))
ordered_back_all <- ordered_back_all$result
ordered_back_all <- ordered_back_all[ordered_back_all$term_size > 15 &
 ordered_back_all$term_size < 2000 & ordered_back_all$intersection_size > 2,]
ordered_back_all_tf <- gprofiler2::gost(query = rowname$rowname[1:150], organism = "mmusculus",
 ordered_query = TRUE, significant = TRUE, exclude_iea = FALSE, multi_query = FALSE,
 measure_underrepresentation = FALSE, evcodes = FALSE, user_threshold = 0.05,
  correction_method = "fdr",  numeric_ns = "", sources = c("TF"))
ordered_back_all_tf <- ordered_back_all_tf$result
ordered_back_all_tf <- ordered_back_all_tf[ordered_back_all_tf$term_size > 15
 & ordered_back_all_tf$term_size < 5000 & ordered_back_all_tf$intersection_size > 2,]
TF = ordered_back_all_tf
BP <- ordered_back_all
bp <- plotBP(BP)
tf <- make_TF_barplot(TF)
```

---

pathway_enrich_internal

Pathway enrichment for cellWeighted_Foldchange's and bulk gene list

---

Description

This function completes pathway enrichment of cellWeighted_Foldchanges and bulk gene list.

Usage

```
pathway_enrich_internal(
  DEGs,
  theSpecies,
  scMappR_vals,
  background_genes,
  output_directory,
  plot_names,
  number_genes = -9,
  newGprofiler = FALSE,
  toSave = FALSE,
  path = NULL
)
```

Arguments

DEGs	Differentially expressed genes (gene_name, padj, log2fc).
theSpecies	Human, mouse, or a charcter that is compatible with gProfileR.
scMappR_vals	cell weighted Fold-changes of differentially expressed genes.
background_genes	A list of background genes to test against.
output_directory	Path to the directory where files will be saved.
plot_names	Names of output.
number_genes	Number of genes to if there are many, many DEGs.
newGprofiler	Whether to use gProfileR or gprofiler2 (T/F).
toSave	Allow scMappR to write files in the current directory (T/F).
path	If toSave == TRUE, path to the directory where files will be saved.

Details

Internal: Pathway analysis of differentially expressed genes (DEGs) and cell weighted Fold-changes (cellWeighted_Foldchanges) for each cell-type. Returns .RData objects of differential analysis as well as plots of the top bulk pathways. It is a wrapper for making barplots, bulk pathway analysis, and gProfiler_cellWeighted_Foldchange.

Value

List with the following elements:

BPs	Enriched biological pathways for each cell-type.
TFs	Enriched transcription factors for each cell-type.

Examples

```
data(PBMC_example)
bulk_DE_cors <- PBMC_example$bulk_DE_cors
bulk_normalized <- PBMC_example$bulk_normalized
odds_ratio_in <- PBMC_example$odds_ratio_in
case_grep <- "_female"
control_grep <- "_male"
max_proportion_change <- 10
print_plots <- FALSE
theSpecies <- "human"
toOut <- scMappR_and_pathway_analysis(bulk_normalized, odds_ratio_in,
                                      bulk_DE_cors, case_grep = case_grep,
                                      control_grep = control_grep, rda_path = "",
                                      max_proportion_change = 10, print_plots = TRUE,
                                       plot_names = "tst1", theSpecies = "human",
                                       output_directory = "tester",
                                   sig_matrix_size = 3000, up_and_downregulated = FALSE,
                                       internet = FALSE)
```

`PBMC_example` *PBMC_scMappR*

Description

Data for scMappR_and_pathway_analysis example.

Usage

```
data(PBMC_example)
```

Format

A list containing three data frames, normalized count data, a signature matrix, and a list of differentially expressed genes.

bulk_normalized A 3231 x 9 matrix where rows are genes, columns are samples, and the matrix is filled with CPM normalized counts.

odds_ratio_in A 2336 x 7 matrix where rows are genes, columns are cell-types and matrix is filled with the odds-ratio that a gene is in each cell-type.

bulk_DE_cors A 59 x 3 matrix of sex-specific genes found between male and female PBMC samples (female biased = upregulated). row and rownames are genes, columns are gene name, FDR adjusted p-value, and log2 fold-change. DEGs were computed with DESeq2 and genes with a log2FC > 1 were kept.

Details

A named list called "PBMC_example" containing the count data, signature matrix, and DEGs. The count data and signature matrix are shortened to fit the size of the package and do not reflect the paper.

Examples

```
data(PBMC_example)
```

---

plotBP *Plot gProfileR Barplot*

Description

Make a barplot of the top biological factors enriched by gProfileR.

Usage

```
plotBP(ordered_back_all, top_bp = 10)
```

Arguments

ordered_back_all	Output of the gProfileR function.
top_bp	The number of pathways you want to plot.

Details

This function takes a gProfileR output and prints the top "top_bp" most significantly enriched p-values before plotting the rank of their p-values.

Value

plotBP A barplot of the number of "top_bp" pathways, ranked by -log10(Pfdr).

Examples

```
data(POA_example)
 POA_generes <- POA_example$POA_generes
 POA_OR_signature <- POA_example$POA_OR_signature
 POA_Rank_signature <- POA_example$POA_Rank_signature
Signature <- as.data.frame(POA_Rank_signature)
rowname <- get_gene_symbol(Signature)
rownames(Signature) <- rowname$rowname
ordered_back_all <- gprofiler2::gost(query = rowname$rowname[1:100], organism = "mmusculus",
 ordered_query = TRUE, significant = TRUE, exclude_iea = FALSE, multi_query = FALSE,
```

```
  measure_underrepresentation = FALSE, evcodes = FALSE, user_threshold = 0.05,
   correction_method = "fdr",   numeric_ns = "", sources = c("GO:BP", "KEGG", "REAC"))
ordered_back_all <- ordered_back_all$result
ordered_back_all <- ordered_back_all[ordered_back_all$term_size > 15
 & ordered_back_all$term_size < 2000 & ordered_back_all$intersection_size > 2,]
ordered_back_all_tf <- gprofiler2::gost(query = rowname$rowname[1:150], organism = "mmusculus",
 ordered_query = TRUE, significant = TRUE, exclude_iea = FALSE, multi_query = FALSE,
  measure_underrepresentation = FALSE, evcodes = FALSE, user_threshold = 0.05,
   correction_method = "fdr",  numeric_ns = "", sources = c("TF"))
ordered_back_all_tf <- ordered_back_all_tf$result
ordered_back_all_tf <- ordered_back_all_tf[ordered_back_all_tf$term_size > 15
& ordered_back_all_tf$term_size < 5000 & ordered_back_all_tf$intersection_size > 2,]
TF = ordered_back_all_tf
BP <- ordered_back_all
bp <- plotBP(BP)
tf <- make_TF_barplot(TF)
```

POA_example *Preoptic_Area*

Description

Data for tissue_scMappR_custom, tissue_scMappR_internal, generes_to_heatmap.

Usage

```
data(POA_example)
```

Format

A list containing three objects: summary statistics of cell-type markers, a signature matrix of odds ratios, and a signature matrix of ranks.

POA_generes A list of 27 data frames containing (up to 30) cell-type markers. Each element of the list is a dataframe where rows are genes, and columns are p-value, log2FC, percentage of cells expressing gene in cell-type, percentage of cells expressing gene in other cell-types, and FDR adjusted p-value.

POA_OR_signature A 266 x 27 matrix where rows are genes, columns are cell-types and matrix is filled with the odds-ratio that a gene is in each cell-type.

POA_Rank_signature A 266 x 27 matrix of matrix where rows are genes, columns are cell-types and matrix is filled with the rank := -log10(P_fdr) that a gene is in each cell-type.

Details

A named list called POA_example (pre-optic area example) containing three objects, POA_generes: a list of truncated dataframes containing summary statistics for each cell-type marker, POA_OR_signature a truncated signature matrix of odds ratio's for cell-types in the POA, and POA_Rank_signature a truncated signature matrix of -log10(Padj) for cell-type markers in the POA.

Examples

```
data(POA_example)
```

---

process_dgTMatrix_lists *Count Matrix To Signature Matrix*

Description

This function takes a list of count matrices, processes them, calls cell-types, and generates signature matrices.

Usage

```
process_dgTMatrix_lists(
  dgTMatrix_list,
  name,
  species_name,
  naming_preference = -9,
  rda_path = "",
  panglao_set = FALSE,
  haveUMAP = FALSE,
  saveSCObject = FALSE,
  internal = FALSE,
  toSave = FALSE,
  path = NULL,
  use_sctransform = FALSE,
  test_ctname = "wilcox",
  genes_integrate = 2000,
  genes_include = FALSE
)
```

Arguments

`dgTMatrix_list`	A list of matrices in the class of dgTMatrix object – sparce object – compatible with Seurat rownames should be of the same species for each.
`name`	The name of the outputted signature matrices, cell-type preferences, and seurat objects if you choose to save them.
`species_name`	Mouse or human symbols, -9 if internal as panglao objects have gene symbol and ensembl strapped together.
`naming_preference`	For cell-type naming, see if cell-types given the inputted tissues are more likely to be named within one of the categories of get_naming_preference_options().
`rda_path`	If saved, directory to where data from scMappR_data is downloaded.
`panglao_set`	If the inputted matrices are from Panglao (i.e. if they're internal).

`haveUMAP`	Save the UMAPs – only use if the package is downloaded with pip.
`saveSCObject`	Save the Seurat object as an RData object (T/F).
`internal`	Was this used as part of the internal processing of Panglao datasets (T/F).
`toSave`	Allow scMappR to write files in the current directory (T/F)
`path`	If toSave == TRUE, path to the directory where files will be saved.
`use_sctransform`	If you should use sctransform or the Normalize/VariableFeatures/ScaleData pipeline (T/F).
`test_ctname`	statistical test for calling CT markers – must be in Seurat
`genes_integrate`	The number of genes to include in the integration anchors feature when combining datasets – passed into process_from_count
`genes_include`	TRUE or FALSE – include 2000 genes in signature matrix or all matrix.

Details

This function is a one line wrapper to process count matrices into a signature matrix. It combines process_from_count, two methods of identifying cell-type identities (GSVA and Fisher's test). Then, it takes the output of cell-type markers and converts it into a signature matrix of p-value ranks and odds ratios. Along the way, it saves the Seurat object (if chosen with saveSCObject), cell-type identites from GSVA (its own obect), and the signature matrices. Cell-type marker outputs are also saved in the generes.RData list. Names of the generes objects and the signature matrices are kept.

Value

List with the following elements:

`wilcoxon_rank_mat_t`	A dataframe containing the signature matrix of ranks (-log10(Padj) * sign(fold-change)).
`wilcoxon_rank_mat_or`	A dataframe containing the signature matrix of odds-ratios.
`generes`	All cell-type markers for each cell-type with p-value and fold changes.

Examples

```
data(sm)
toProcess <- list(example = sm)
tst1 <- process_dgTMatrix_lists(toProcess, name = "testPropcess", species_name = -9,
 naming_preference = "eye", rda_path = "")
```

process_from_count *Count Matrix To Seurat Object*

Description

This function processes a list of count matrices (same species/gene symbols in each list) and converts them to a Seurat object.

Usage

```
process_from_count(
  countmat_list,
  name,
  theSpecies = -9,
  haveUmap = FALSE,
  saveALL = FALSE,
  panglao_set = FALSE,
  toSave = FALSE,
  path = NULL,
  use_sctransform = FALSE,
  genes_integrate = 2000,
  genes_include = FALSE
)
```

Arguments

countmat_list	A list of count matrices that will be be integrated using the IntegrationAnchors features they should have the same rownames.
name	The output of the normalzied and fused Suerat object if you choose to keep it.
theSpecies	Gene symbols for human, mouse, or -9 if internal. If your species is not human or mouse gene symbols, make sure that you have "MT-" before your mitochondrial gene names then pick "human".
haveUmap	Write a UMAP (T/F).
saveALL	Save the Seurat object generated (T/F).
panglao_set	If the function is being used from internal (T/F).
toSave	Allows scMappR to print files and make directories locally (T/F).
path	If toSave == TRUE, path to the directory where files will be saved.
use_sctransform	If you should use sctransform or the Normalize/VariableFeatures/ScaleData pipeline (T/F).
genes_integrate	The number of genes to include in the integration anchors feature when combining datasets
genes_include	TRUE or FALSE – include 2000 genes in signature matrix or all matrix.

Details

This function takes a list of count matrices and returns a Seurat object of the count matrices integrated using Seurat v3 (with sctransform and IntegrationAnchors). Different options are used when the function is being ran internally (i.e. reprocessing count matrices from PanglaoDB) or if it is running from custom scRNA-seq data. For larger scRNA-seq datasets (~20k + cells), it is likely that this function will be required to run on an hpc.

Value

`process_from_count` A processed and integrated Seurat object that has been scaled and clustered. It can be returned as an internal object or also stored as an RData object if neccesary.

Examples

```
data(sm)
toProcess <- list(example = sm)
tst1 <- process_from_count(toProcess, "testProcess")
```

---

scMappR_and_pathway_analysis

Generate cellWeighted_Foldchange, visualize, and enrich.

Description

This function generates cell weighted Fold-changes (cellWeighted_Foldchange), visualizes them in a heatmap, and completes pathway enrichment of cellWeighted_Foldchanges and bulk gene list.

Usage

```
scMappR_and_pathway_analysis(
  count_file,
  signature_matrix,
  DEG_list,
  case_grep,
  control_grep,
  rda_path = "",
  max_proportion_change = -9,
  print_plots = T,
  plot_names = "scMappR",
  theSpecies = "human",
  output_directory = "scMappR_analysis",
  sig_matrix_size = 3000,
```

```
  drop_unknown_celltype = TRUE,
  internet = TRUE,
  up_and_downregulated = FALSE,
  gene_label_size = 0.4,
  number_genes = -9,
  toSave = FALSE,
  newGprofiler = FALSE,
  path = NULL
)
```

Arguments

count_file
: Normalized RNA-seq count matrix where rows are gene symbols and columns are individuals. Either the object tself of the path of a .tsv file.

signature_matrix
: Signature matrix (recommended odds ratios) of cell-type specificity of genes. Either the object itself or a pathway to a .RData file containing an object named "wilcoxon_rank_mat_or" – generally internal.

DEG_list
: An object with the first column as gene symbols within the bulk dataset (doesn't have to be in signature matrix), second column is the adjusted p-value, and the third the log2FC path to a .tsv file containing this info is also acceptable.

case_grep
: Tag in the column name for cases (i.e. samples representing upregulated) OR an index of cases.

control_grep
: Tag in the column name for controls (i.e. samples representing downregulated OR an index of controls).

rda_path
: If downloaded, path to where data from scMappR_data is stored.

max_proportion_change
: Maximum cell-type proportion change – may be useful if there are many rare cell-types.

print_plots
: Whether boxplots of the estimated CT proportion for the leave-one-out method of CT deconvolution should be printed. The same name of the plots will be completed for top pathways.

plot_names
: The prefix of plot pdf files.

theSpecies
: -9 if using a pre-computed count matrix from scMappR, human, mouse, or a specied directly compatible with gProfileR. Removes Ensembl symbols if appended.

output_directory
: The name of the directory that will contain output of the analysis.

sig_matrix_size
: Number of genes in signature matrix for cell-type deconvolution.

drop_unknown_celltype
: Whether or not to remove "unknown" cell-types from the signature matrix.

internet
: Whether you have stable Wifi (T/F).

up_and_downregulated
: Whether you are additionally splitting up/downregulated genes (T/F).

gene_label_size
: The size of the gene label on the plot.

number_genes
: The number of genes to cut-off for pathway analysis (good with many DEGs).

toSave
: Allow scMappR to write files in the current directory (T/F).

newGprofiler
: Whether to use gProfileR or gprofiler2 (T/F).

path
: If toSave == TRUE, path to the directory where files will be saved.

Details

This function generates cellWeighted_Foldchanges for every cell-type (see deconvolute_and_contextualize), as well as the relative cell-type proportions (which will be reutrned and pushed through). Then, it generates heatmaps of all cellWeighted_Foldchanges, cellWeighted_Foldchanges overlapping with the signature matrix, the entire signature matrix, the cell-type preference values from the signature matrix that overlap with inputted differentially expressed genes. Then, if you have Wifi, it will complete gProfileR of the reordered cellWeighted_Foldchanges as well as a the ordered list of genes. This function is a wrapper for deconvolute_and_contextualize and pathway_enrich_internal.

Value

List with the following elements:

cellWeighted_Foldchanges
: Cellweighted Fold-changes for all differentially expressed genes.

paths
: Enriched biological pathways for each cell-type.

TFs
: Enirched TFs for each cell-type.

Examples

```
data(PBMC_example)
bulk_DE_cors <- PBMC_example$bulk_DE_cors
bulk_normalized <- PBMC_example$bulk_normalized
odds_ratio_in <- PBMC_example$odds_ratio_in
case_grep <- "_female"
control_grep <- "_male"
max_proportion_change <- 10
print_plots <- FALSE
theSpecies <- "human"
toOut <- scMappR_and_pathway_analysis(bulk_normalized, odds_ratio_in,
                                      bulk_DE_cors, case_grep = case_grep,
                                      control_grep = control_grep, rda_path = "",
                                      max_proportion_change = 10, print_plots = TRUE,
                                       plot_names = "tst1", theSpecies = "human",
                                       output_directory = "tester",
                                   sig_matrix_size = 3000, up_and_downregulated = FALSE,
                                       internet = FALSE)
```

scMappR_tissues *scMappR_tissues*

Description

Tissues available in scMappR as of February 2020.

Usage

```
data(scMappR_tissues)
```

Format

A vector of tissue names available for tissue_scMappR_internal or to download and use in scMappR_and_pathway_analysis.

scMappR_tissues A list of 174 tissue names from PanglaoDB.

Details

A vector of tissues available in scMappR as of February 2020.

Examples

```
data(scMappR_tissues)
```

seurat_to_generes *Identify all cell-type markers*

Description

Takes processed Seurat matrix and identifies cell-type markers with FindMarkers.

Usage

```
seurat_to_generes(pbmc, test = "wilcox")
```

Arguments

pbmc	Processed Seurat object.
test	statistical test for calling CT markers – must be in Seurat

Details

Internal: This function runs the FindMarkers function from seurat in a loop, will use the Seurat v2 or Seurat v3 object after identifying which Seurat object is inputted. It then takes the output of the FindMarkers and puts it in a list, returning it.

Value

`seurat_to_generes` A list of genes where their over-representation in the i'th cell-type is computed. Each element contains the gene name, adjusted p-value, and the log2Fold-Change of each gene being present in that cell-type.

Examples

```
data(sm)
toProcess <- list(example = sm)
tst1 <- process_from_count(toProcess, "testProcess")
generes <- seurat_to_generes(tst1)
```

`single_gene_preferences` *Single cell-type gene preferences*

Description

Measure enrichment of individual cell-types in a signature matrix.

Internal function as part of tissue_scMappR_internal(). This function takes genes preferentially expressed within a gene list, each cell-type and the background (i.e. all genes within the signature matrix) before completing the cell-type specific enrichment of the inputted gene list on each cell type. This function then returns a table describing the cell-type enrichments (p-value and odds ratio) of each cell-type.

Usage

```
single_gene_preferences(
  hg_short,
  hg_full,
  study_name,
  outDir,
  toSave = FALSE,
  path = NULL
)
```

Arguments

`hg_short`	A list with two objects: a "preferences" and a "genesIn". Preferences is a list of gene symbols over-represented in each cell-type and genesIn were all the inputted genes.
`hg_full`	The same as hg_short but for every gene in the signature matrix.

study_name	Name of output table.
outDir	Directory where table is outputted.
toSave	Allow scMappR to write files in the current directory (T/F).
path	If toSave == TRUE, path to the directory where files will be saved.

Value

single_gene_preferences A gene-set enrichment table of individual cell-type enrichment.

Examples

```
# load in signature matrices
data(POA_example)
POA_generes <- POA_example$POA_generes
POA_OR_signature <- POA_example$POA_OR_signature
POA_Rank_signature <- POA_example$POA_Rank_signature
sig <- get_gene_symbol(POA_Rank_signature)
Signature <- POA_Rank_signature
rownames(Signature) <- sig$rowname
genes <- rownames(Signature)[1:60]
heatmap_test <- tissue_scMappR_custom( genes, signature_matrix = Signature,
                                     output_directory =  "scMappR_test", toSave = FALSE)
single_preferences <- heatmap_test$single_celltype_preferences
```

---

sm *single_cell_process*

Description

Example data for processing scRNA-seq count data with Seurat.

Usage

```
data(sm)
```

Format

A 752 x 236 matrix of class dgCMatrix where rows are genes and columns are cells. Data matrix is filled with counts detected from scRNAseq

TCTCTAACACAGGCCT Barcode of one of the sequenced cells present. Each column is the count from a scRNA-seq dataset reprocessed by PanglaoDB.

Details

A dgCMatrix object containing count data for scRNA-seq processing.

Examples

```
data(sm)
```

---

tissue_by_celltype_enrichment

tissue_by_celltype_enrichment

Description

This function uses a Fisher's-exact-test to rank gene-set enrichment.

Usage

```
tissue_by_celltype_enrichment(
  gene_list,
  species,
  name = "CT_Tissue_example",
  p_thresh = 0.05,
  rda_path = "",
  isect_size = 3,
  return_gmt = FALSE
)
```

Arguments

gene_list	A character vector of gene symbols with the same designation (e.g. mouse symbol - mouse, human symbol - human) as the gene set database.
species	Species of cell-type marker to use ('human' or 'mouse').
name	Name of the pdf to be printed.
p_thresh	The Fisher's test cut-off for a cell-marker to be enriched.
rda_path	Path to a .rda file containing an object called "gmt". Either human or mouse cell-type markers split by experiment. If the correct file isn't present they will be downloaded from https://github.com/wilsonlabgroup/scMappR_Data.
isect_size	Number of genes in your list and the cell-type.
return_gmt	Return .gmt file – reccomended if downloading from online as it may have updated (T/F).

Details

Complete a Fisher's exact test of an input list of genes against one of the two curated tissue by cell-type marker datasets from scMappR.

Value

List with the following elements:

enriched	Data frame of enriched cell-types from tissues.
gmt	Cell-markers in enriched cell-types from tissues.

Examples

```
data(POA_example)
POA_generes <- POA_example$POA_generes
POA_OR_signature <- POA_example$POA_OR_signature
POA_Rank_signature <- POA_example$POA_Rank_signature
Signature <- POA_Rank_signature
rowname <- get_gene_symbol(Signature)
rownames(Signature) <- rowname$rowname
genes <- rownames(Signature)[1:100]

enriched <- tissue_by_celltype_enrichment(gene_list = genes,
species = "mouse",p_thresh = 0.05, isect_size = 3)
```

---

tissue_scMappR_custom *Gene List Visualization and Enrichment with Custom Signature Matrix*

Description

This function visualizes signature matrix, clusters subsetted genes, completes enrichment of individual cell-types and co-enrichment.

Usage

```
tissue_scMappR_custom(
  gene_list,
  signature_matrix,
  output_directory = "custom_test",
  toSave = FALSE,
  path = NULL,
  gene_cutoff = 1,
  is_pvalue = TRUE
)
```

Arguments

- `gene_list` A list of gene symbols matching that of the signature_matrix. Any gene symbol is acceptable.
- `signature_matrix` Pre=computed signature matrix with matching gene names.
- `output_directory` Directory made containing output of functions.
- `toSave` Allow scMappR to write files in the current directory (T/F).
- `path` If toSave == TRUE, path to the directory where files will be saved.
- `gene_cutoff` Value cut-off (generally rank := log10(Padj)) for a gene to be considered a marker.
- `is_pvalue` If signature matrix is p-value before rank is applied (not recommended) (T/F).

Details

This function is roughly the same as tissue_scMappR_internal, however now there is a custom signature matrix. It generates a heatmap of the signature matrix and your inputted gene list, as well as single cell-type and co-celltype enrichment.

Value

List with the following elements:

- `background_heatmap` Data frame of the entire gene by cell-type signature matrix inputted.
- `gene_list_heatmap` Data frame of inputted signature matrix subsetted by input genes.
- `single_celltype_preferences` Data frame of enriched cell-types.
- `group_celtype_preference` Data frame of groups of cell-types enriched by the same genes.

Examples

```
# load in signature matrices
data(POA_example)
POA_generes <- POA_example$POA_generes
POA_OR_signature <- POA_example$POA_OR_signature
POA_Rank_signature <- POA_example$POA_Rank_signature
sig <- get_gene_symbol(POA_Rank_signature)
Signature <- POA_Rank_signature
rownames(Signature) <- sig$rowname
genes <- rownames(Signature)[1:60]
heatmap_test <- tissue_scMappR_custom( genes, signature_matrix = Signature,
                                     output_directory =  "scMappR_test", toSave = FALSE)
```

tissue_scMappR_internal

Gene List Visualization and Enrichment (Internal)

Description

This function loops through every signature matrix in a particular tissue and generates heatmaps, cell-type preferences, and co-enrichment.

Usage

```
tissue_scMappR_internal(
  gene_list,
  species,
  output_directory,
  tissue,
  rda_path = "",
  cluster = "Pval",
  genecex = 0.01,
  raw_pval = FALSE,
  path = NULL,
  toSave = FALSE,
  drop_unkown_celltype = FALSE
)
```

Arguments

`gene_list`	A list of gene symbols, mouse or human.
`species`	"mouse", "human" or "-9" if using a precomputed signature matrix.
`output_directory`	If toSave = TRUE, the name of the output directory that would be built.
`tissue`	Name of the tissue in "get_tissues".
`rda_path`	Path to the .rda file containing all of the signature matrices.
`cluster`	'Pval' or 'OR' depending on if you wantto cluster odds ratios or pvalues of CT preferences.
`genecex`	The size of the gene names of the rows in the heatmap.
`raw_pval`	If the inputted signature matrix are raw (untransformed) p-values – recommended to generate rank first (T/F).
`path`	If toSave == TRUE, path to the directory where files will be saved.
`toSave`	Allow scMappR to write files in the current directory (T/F).
`drop_unkown_celltype`	Whether or not to remove "unknown" cell-types from the signature matrix (T/F).

Details

This function takes a list of genes and a tissue that is contained in current signature matrices before and generating heatmaps of cell-type preferences. It then completes cell-type enrichment of each individual cell-type, then, if more than two cell-types are signficiantly enriched, co-enrichemnt of those enriched cell-types is then computed.

Value

List with the following elements:

`background_heatmap`
: Data frame of the entire gene by cell-type signature matrix inputted.

`gene_list_heatmap`
: Data frame of inputted signature matrix subsetted by input genes.

`single_celltype_preferences`
: Data frame of enriched cell-types.

`group_celtype_preference`
: Data frame of groups of cell-types enriched by the same genes.

Examples

```
data(POA_example) # region to preoptic area
Signature <- POA_example$POA_Rank_signature # signature matrix
rowname <- get_gene_symbol(Signature) # get signature
rownames(Signature) <- rowname$rowname
genes <- rownames(Signature)[1:60]
rda_path1 = "" # data directory (if it exists)

# set toSave = TRUE and path = output directory of your choice
internal <- tissue_scMappR_internal(genes, "mouse", output_directory = "scMappR_TesInternal",
                                    tissue = "hypothalamus", toSave = FALSE)
```

`tochr` *To Character.*

Description

This function checks if your vector is not a character and if not, will convert it to a character.

Usage

```
tochr(x)
```

Arguments

`x`	A character, factor or numeric vector.

Value

`tochr` Returns a character vector.

Examples

```
 # vector of factors
 fact <- factor(c("a", "b", "c", "d"))
 # convert to character
 char <- tochr(fact)
```

---

toNum *To Numeric.*

Description

This function checks if your vector is not a character and if it is, then converts it to a numeric.

Usage

```
toNum(x)
```

Arguments

`x`	A character, factor, or numeric vector.

Value

`toNum` Returns a numeric vector.

Examples

```
 # vector of factors
fact <- factor(c("1", "2", "3", "4"))
# convert to numeric
num <- toNum(fact)
```

topgenes_extract *Extract Top Markers*

Description

Internal – Extracts strongest cell-type markers from a Seurat object.

Usage

```
topgenes_extract(generes, padj = 0.05, FC = 1.5, topNum = 30)
```

Arguments

generes	A list of cell-tpe markers with fold-changes and p-vlaues (FindMarkers output in Seurat).
padj	The p-value (FDR) cutoff.
FC	The fold-change cutoff.
topNum	The number of genes to extract.

Details

Internal, this function runs through a list of outputs from FindMarkers objects in Seurat and will extract genes past a padj and fold-change threshold. Then it extracts the topNum number of genes. if you have not used the FindMarkers function, then a list of summary statistics with fold-change designated by avg_logFC and p-val by p_val_adj.

Value

`topgenes_extract` Returns a list of character vectors with the top (topNum) of gene markers for each cell-type.

Examples

```
# load generes object
data(POA_example)
topGenes <- topgenes_extract(POA_example$POA_generes)
```

two_method_pathway_enrichment
two_method_pathway_enrichment

Description

Pathway analysis of each cell-type based on cell-type specificity and rank improvement by scMappR.

Usage

```
two_method_pathway_enrichment(
  DEGs,
  theSpecies,
  scMappR_vals,
  background_genes = NULL,
  output_directory = "output",
  plot_names = "reweighted",
  number_genes = -9,
  newGprofiler = FALSE,
  toSave = FALSE,
  path = NULL
)
```

Arguments

- `DEGs` Differentially expressed genes (gene_name, padj, log2fc).
- `theSpecies` Human, mouse, or a charcter that is compatible with gProfileR.
- `scMappR_vals` cell weighted Fold-changes of differentially expressed genes.
- `background_genes` A list of background genes to test against. NULL assumes all genes in gprofiler gene set databases.
- `output_directory` Path to the directory where files will be saved.
- `plot_names` Names of output.
- `number_genes` Number of genes to if there are many, many DEGs.
- `newGprofiler` Whether to use gProfileR or gprofiler2 (T/F).
- `toSave` Allow scMappR to write files in the current directory (T/F).
- `path` If toSave == TRUE, path to the directory where files will be saved.

Details

This function re-ranks cwFoldChanges based on their absolute ct specificity scores (per-celltype) as well as their rank increase in cell-type specificity before completing an ordered pathway analysis. In the second method, only genes with a rank increase in cell-type specificity were included

Value

List with the following elements:

`rank_increase` A list containing the degree of rank change between bulk DE genes and cwFold-changes. Pathway enrichment and tf enrichment of these reranked genes.

`non_rank_increase` list of DFs containing the pathway and TF enrichment of cwFold-changes.

Examples

```
# load data for scMappR
data(PBMC_example)
bulk_DE_cors <- PBMC_example$bulk_DE_cors
bulk_normalized <- PBMC_example$bulk_normalized
odds_ratio_in <- PBMC_example$odds_ratio_in
case_grep <- "_female"
control_grep <- "_male"
max_proportion_change <- 10
print_plots <- FALSE
theSpecies <- "human"

# calculate cwFold-changes
toOut <- scMappR_and_pathway_analysis(bulk_normalized, odds_ratio_in,
                                      bulk_DE_cors, case_grep = case_grep,
                                      control_grep = control_grep, rda_path = "",
                                      max_proportion_change = 10, print_plots = TRUE,
                                      plot_names = "tst1", theSpecies = "human",
                                      output_directory = "tester",
                                   sig_matrix_size = 3000, up_and_downregulated = FALSE,
                                      internet = FALSE)

# complete pathway enrichment using both methods
twoOutFiles <- two_method_pathway_enrichment(bulk_DE_cors, "human",
scMappR_vals = toOut$cellWeighted_Foldchange, background_genes = rownames(bulk_normalized),
output_directory = "newfun_test",plot_names = "nonreranked_", toSave = FALSE)
```

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